

PATENT APPLICATION

**METHODS OF PREDICTING
MUSCULOSKELETAL DISEASE**

Inventors: Siau-Way LIEW
 Philipp LANG
 Daniel STEINES

Assignee: Imaging Therapeutics, Inc.

Entity: Small

Robins & Pasternak LLP
1731 Embarcadero Road, Suite 230
Palo Alto, CA 94303
Tel: 650 493-3400
Fax: 650 493-3440

METHODS OF PREDICTING MUSCULOSKELETAL DISEASE

By

**Siau-Way Liew
Philipp Lang
Daniel Steines**

5

CROSS REFERENCE TO RELATED APPLICATIONS

[0001] This application is a continuation-in-part of U.S. Serial No. 10/665,725,
10 filed September 16, 2003, which in turn claims the benefit of U.S. Provisional
Patent Application Serial No. 60/411,413, filed on September 16, 2002 and also
claims the benefit of U.S. Provisional Patent Application Serial No. 60/438,641,
filed on January 7, 2003, from which applications priority is hereby claimed under
35 USC §§ 119/120, and which applications are hereby incorporated herein by
15 reference in their entireties herein.

TECHNICAL FIELD

[0002] This invention relates to using imaging methods for diagnosis,
prognostication, monitoring and management of disease, particularly where that
disease affects the musculoskeletal system. This invention identifies novel
20 imaging markers for use in diagnosis, prognostication, monitoring and
management of disease, including musculoskeletal disease.

BACKGROUND

[0003] Osteoporosis and osteoarthritis are among the most common
conditions to affect the musculoskeletal system, as well as frequent causes of
25 locomotor pain and disability. Osteoporosis can occur in both human and animal
subjects (e.g. horses). Osteoporosis (OP) and osteoarthritis (OA) occur in a
substantial portion of the human population over the age of fifty. The National
Osteoporosis Foundation estimates that as many as 44 million Americans are

affected by osteoporosis and low bone mass. In 1997 the estimated cost for osteoporosis related fractures was \$13 billion. That figure increased to \$17 billion in 2002 and is projected to increase to \$210-240 billion by 2040. Currently it is expected that one in two women over the age of 50 will suffer an osteoporosis-related fracture.

[0004] Imaging techniques are important diagnostic tools, particularly for bone related conditions such as OP and OA. Currently available techniques for the noninvasive assessment of the skeleton for the diagnosis of osteoporosis or the evaluation of an increased risk of fracture include dual x-ray absorptiometry (DXA) (Eastell et al. (1998) *New Engl J. Med* 338:736-746); quantitative computed tomography (QCT) (Cann (1988) *Radiology* 166:509-522); peripheral DXA (pDXA) (Patel et al. (1999) *J Clin Densitom* 2:397-401); peripheral QCT (pQCT) (Gluer et. al. (1997) *Semin Nucl Med* 27:229-247); x-ray image absorptiometry (RA) (Gluer et. al. (1997) *Semin Nucl Med* 27:229-247); and quantitative ultrasound (QUS) (Njeh et al. "Quantitative Ultrasound: Assessment of Osteoporosis and Bone Status" 1999, Martin-Dunitz, London England; U.S. Patent No. 6,077,224, incorporated herein by reference in its entirety). (See, also, WO 9945845; WO 99/08597; and U.S. Patent No. 6,246,745).

[0005] DXA of the spine and hip has established itself as the most widely used method of measuring BMD. Tothill, P. and D.W. Pye, (1992) *Br J Radiol* 65:807-813. The fundamental principle behind DXA is the measurement of the transmission through the body of x-rays of 2 different photon energy levels. Because of the dependence of the attenuation coefficient on the atomic number and photon energy, measurement of the transmission factors at 2 energy levels enables the area densities (i.e., the mass per unit projected area) of 2 different types of tissue to be inferred. In DXA scans, these are taken to be bone mineral (hydroxyapatite) and soft tissue, respectively. However, it is widely recognized that the accuracy of DXA scans is limited by the variable composition of soft tissue. Because of its higher hydrogen content, the attenuation coefficient of fat is

different from that of lean tissue. Differences in the soft tissue composition in the path of the x-ray beam through bone compared with the adjacent soft tissue reference area cause errors in the BMD measurements, according to the results of several studies. Tothill, P. and D.W. Pye, (1992) *Br J Radiol*, 65:807-813; Svendsen, O.L., et al., (1995) *J Bone Min Res* 10:868-873. Moreover, DXA systems are large and expensive, ranging in price between \$75,000 and \$150,000.

[0006] Quantitative computed tomography (QCT) is usually applied to measure the trabecular bone in the vertebral bodies. Cann (1988) *Radiology* 166:509-522. QCT studies are generally performed using a single kV setting (single-energy QCT), when the principal source of error is the variable composition of the bone marrow. However, a dual-kV scan (dual-energy QCT) is also possible. This reduces the accuracy errors but at the price of poorer precision and higher radiation dose. Like DXA, however, QCT are very expensive and the use of such equipment is currently limited to few research centers.

[0007] Quantitative ultrasound (QUS) is a technique for measuring the peripheral skeleton. Njeh et al. (1997) *Osteoporosis Int* 7:7-22; Njeh et al. Quantitative Ultrasound: Assessment of Osteoporosis and Bone Status. 1999, London, England: Martin Dunitz. There is a wide variety of equipment available, with most devices using the heel as the measurement site. A sonographic pulse passing through bone is strongly attenuated as the signal is scattered and absorbed by trabeculae. Attenuation increases linearly with frequency, and the slope of the relationship is referred to as broadband ultrasonic attenuation (BUA; units: dB/MHz). BUA is reduced in patients with osteoporosis because there are fewer trabeculae in the calcaneus to attenuate the signal. In addition to BUA, most QUS systems also measure the speed of sound (SOS) in the heel by dividing the distance between the sonographic transducers by the propagation time (units: m/s). SOS values are reduced in patients with osteoporosis because

with the loss of mineralized bone, the elastic modulus of the bone is decreased. There remain, however, several limitations to QUS measurements. The success of QUS in predicting fracture risk in younger patients remains uncertain. Another difficulty with QUS measurements is that they are not readily encompassed
5 within the WHO definitions of osteoporosis and osteopenia. Moreover, no intervention thresholds have been developed. Thus, measurements cannot be used for therapeutic decision-making.

[0008] There are also several technical limitations to QUS. Many devices use a foot support that positions the patient's heel between fixed transducers. Thus,
10 the measurement site is not readily adapted to different sizes and shapes of the calcaneus, and the exact anatomic site of the measurement varies from patient to patient. It is generally agreed that the relatively poor precision of QUS measurements makes most devices unsuitable for monitoring patients' response to treatment. Gluer (1997) *J Bone Min Res* 12:1280-1288.

[0009] Radiographic absorptiometry (RA) is a technique that was developed
15 many years ago for assessing bone density in the hand, but the technique has recently attracted renewed interest. Gluer et al. (1997) *Semin Nucl Med* 27:229-247. With this technique, BMD is measured in the phalanges. The principal disadvantage of RA of the hand is the relative lack of high turnover trabecular
20 bone. For this reason, RA of the hand has limited sensitivity in detecting osteoporosis and is not very useful for monitoring therapy-induced changes.

[0010] Peripheral x-ray absorptiometry methods such as those described above are substantially cheaper than DXA and QCT with system prices ranging between \$15,000 and \$35,000. However, epidemiologic studies have shown that
25 the discriminatory ability of peripheral BMD measurements to predict spine and hip fractures is lower than when spine and hip BMD measurements are used. Cummings et al. (1993) *Lancet* 341:72-75; Marshall et al. (1996) *Br Med J* 312:1254-1259. The main reason for this is the lack of trabecular bone at the

measurement sites used with these techniques. In addition, changes in forearm or hand BMD in response to hormone replacement therapy, bisphosphonates, and selective estrogen receptor modulators are relatively small, making such measurements less suitable than measurements of principally trabecular bone for monitoring response to treatment. Faulkner (1998) *J Clin Densitom* 1:279-285; Hoskings et al. (1998) *N Engl J Med* 338:485-492. Although attempts to obtain information on bone mineral density from dental x-rays have been attempted (See, e.g., Shrout et al. (2000) *J. Periodonol.* 71:335-340; Verhoeven et al. (1998) *Clin Oral Implants Res* 9(5):333-342), these have not provided accurate and reliable results.

[0011] Furthermore, current methods and devices do not generally take into account bone structure analyses. See, e.g., Ruttimann et al. (1992) *Oral Surg Oral Med Oral Pathol* 74:98-110; Southard & Southard (1992) *Oral Surg Oral Med Oral Pathol* 73:751-9; White & Rudolph, (1999) *Oral Surg Oral Med Oral Pathol Oral Radiol Endod* 88:628-35.

[0012] The present invention discloses novel methods and techniques for predicting musculoskeletal disease, particularly methods and compositions that result in the ability to obtain accurate predictions about disease based on bone mineral density and/or bone structure information obtained from images (e.g., radiographic images) and data.

SUMMARY OF THE INVENTION

[0013] The invention discloses a method for analyzing at least one of bone mineral density, bone structure and surrounding tissue. The method typically comprises: (a) obtaining an image of a subject; (b) locating a region of interest on the image; (c) obtaining data from the region of interest; and (d) deriving data selected from the group of qualitative and quantitative from the image data obtained at step c.

[0014] A system is also provided for predicting a disease. Any of these systems can include the steps of: (a) obtaining image data of a subject; (b) obtaining data from the image data wherein the data obtained is at least one of quantitative and qualitative data; and (c) comparing the at least one of quantitative and qualitative data in step b to at least one of: a database of at least one of quantitative and qualitative data obtained from a group of subjects; at least one of quantitative and qualitative data obtained from the subject; and at least one of a quantitative and qualitative data obtained from the subject at time T_n .

[0015] In certain aspects, described herein are methods of diagnosing, monitoring and/or predicting bone or articular disease (e.g., the risk of fracture) in a subject, the method comprising the steps of: determining one or more micro-structural parameters, one or more macroanatomical parameters or biomechanical parameters of a joint in said subject; and combining at least two of said parameters to predict the risk of bone or articular disease. The micro-structural, macroanatomical and/or biomechanical parameters may be, for example, one or more of the measurements/parameters shown in Tables 1, 2 and/or 3. In certain embodiments, one or more micro-structural parameters and one or more macro-anatomical parameters are combined. In other embodiments, one or more micro-structural parameters and one or more biomechanical parameters are combined. In further embodiments, one or more macroanatomical parameters and one or more biomechanical parameters are combined. In still further embodiments, one or more macroanatomical parameters, one or more micro-structural parameters and one or more biomechanical parameters are combined.

[0016] In any of the methods described herein, the comparing may be comprise univariate, bivariate and/or multivariate statistical analysis of one or more of the parameters. In certain embodiments, the methods may further

comprise comparing said parameters to data derived from a reference database of known disease parameters.

5 **[0017]** In any of the methods described herein, the parameters are determined from an image obtained from the subject. In certain embodiments, the image comprises one or more regions of bone (*e.g.*, patella, femur, tibia, fibula, pelvis, spine, etc). The image may be automatically or manually divided into two or more regions of interest. Furthermore, in any of the methods described herein, the image may be, for example, an x-ray image, a CT scan, an MRI or the like and optionally includes one or more calibration phantoms.

10 **[0018]** In any of the methods described herein, the predicting includes performing univariate, bivariate or multivariate statistical analysis of the analyzed data and referencing the statistical analysis values to a fracture risk model. Fracture risk models can comprise, for example, data derived from a reference database of known fracture loads with their corresponding values of macro-
15 anatomical, micro-anatomical parameters, and/or clinical risk factors.

20 **[0019]** In another aspect, the invention includes a method of determining the effect of a candidate agent on a subject's prognosis for musculoskeletal disease comprising: predicting a first risk of musculoskeletal disease in subject according to any of the predictive methods described herein; administering a candidate
25 agent to the subject; predicting a second risk of the musculoskeletal disease in the subject according to any of the predictive methods described herein; and comparing the first and second risks, thereby determining the effect of the candidate on the subject's prognosis for the disease. In any of these methods, the candidate agent can be administered to the subject in any modality, for
30 example, by injection (intramuscular, subcutaneous, intravenous), by oral administration (*e.g.*, ingestion), topical administration, mucosal administration or the like. Furthermore, the candidate agent may be a small molecule, a

pharmaceutical, a biopharmaceutical, an agropharmaceuticals and/or combinations thereof.

[0020] In other aspects, the invention includes a kit that is provided for aiding in the prediction of musculoskeletal disease (e.g., fracture risk). The kit typically
5 comprises a software program that uses information obtained from an image to predict the risk or disease (e.g., fracture). The kit can also include a database of measurements for comparison purposes. Additionally, the kit can include a subset of a database of measurements for comparisons.

[0021] In any of these methods, systems or kits, additional steps can be
10 provided. Such additional steps include, for example, enhancing image data.

[0022] Suitable subjects for these steps include for example mammals, humans and horses. Suitable anatomical regions of subjects include, for example, dental, spine, hip, knee and bone core x-rays.

[0023] A variety of systems can be employed to practice the inventions.
15 Typically at least one of the steps of any of the methods is performed on a first computer. Although, it is possible to have an arrangement where at least one of the steps of the method is performed on a first computer and at least one of the steps of the method is performed on a second computer. In this scenario the first computer and the second computer are typically connected. Suitable connections
20 include, for example, a peer to peer network, direct link, intranet, and internet.

[0024] It is important to note that any or all of the steps of the inventions disclosed can be repeated one or more times in series or in parallel with or without the repetition of other steps in the various methods. This includes, for example repeating the step of locating a region of interest, or obtaining image
25 data.

[0025] Data can also be converted from 2D to 3D to 4D and back; or from 2D to 4D. Data conversion can occur at multiple points of processing the information.

For example, data conversion can occur before or after pattern evaluation and/or analysis.

[0026] Any data obtained, extracted or generated under any of the methods can be compared to a database, a subset of a database, or data previously
5 obtained, extracted or generated from the subject. For example, known fracture load can be determined for a variety of subjects and some or all of this database can be used to predict fracture risk by correlating one or more macro-anatomical or structural parameters (Tables 1, 2 and/or 3) with data from a reference database of fracture load for age, sex, race, height and weight matched
10 individuals.

[0027] The present invention provides methods that allow for the analysis of bone mineral density, bone and/or cartilage structure and morphology and/or surrounding tissue from images including electronic images and, accordingly, allows for the evaluation of the effect(s) of an agent (or agents) on bone and/or
15 cartilage. It is important to note that an effect on bone and/or cartilage can occur in agents intended to have an effect, such as a therapeutic effect, on bone and/or cartilage as well as agents intended to primarily effect other tissues in the body but which have a secondary, or tangential, effect on bone and/or cartilage. The images (e.g., x-ray images) can be, for example, dental, hip, spine or other
20 radiographs and can be taken from any mammal. The images can be in electronic format.

[0028] The invention includes a method to derive quantitative information on bone structure and/or bone mineral density from an image comprising (a) obtaining an image, wherein the image optionally includes an external standard
25 for determining bone density and/or structure; and (b) analyzing the image obtained in step (a) to derive quantitative information on bone structure. The image is taken of a region of interest (ROI). Suitable ROI include, for example, a hip radiograph or a dental x-ray obtained on dental x-ray film, including the

mandible, maxilla or one or more teeth. In certain embodiments, the image is obtained digitally, for example using a selenium detector system, a silicon detector system or a computed radiography system. In other embodiments, the image can be digitized from film, or another suitable source, for analysis.

- 5 **[0029]** A method is included where one or more candidate agents can be tested for its effects on bone. Again, the effect can be a primary effect or a secondary effect. For example, images obtained from the subject can be evaluated prior to administration of a candidate agent to predict the risk of disease in the absence of the agent. After administration of the candidate
- 10 agent(s), an electronic image of the same portion of a bone of the subject can be obtained and analyzed as described herein to predict the risk of musculoskeletal disease. The risk of disease prior to administration of the candidate agent and after administration can then be compared to determine if the agent had any effect on disease prognosis. Information on bone structure can relate to a
- 15 variety of parameters, including the parameters shown in Table 1, Table 2 and Table 3, *infra*. The images or data may also be compared to a database of images or data (*e.g.*, “known” images or data). The candidate agent can, for example, be molecules, proteins, peptides, naturally occurring substances, chemically synthesized substances, or combinations and cocktails thereof.
- 20 Typically, an agent includes one or more drugs. Further, the agent can be evaluated for the ability to effect bone diseases such as the risk of bone fracture (*e.g.*, osteoporotic fracture).

- [0030]** In any of the methods described herein, the analysis can comprise using one or more computer programs (or units). Additionally, the analysis can
- 25 comprise identifying one or more regions of interest (ROI) in the image, either prior to, concurrently or after analyzing the image, *e.g.* for information on bone mineral density and/or bone structure. The bone density information can be, for example, areas of highest, lowest or median density. Bone structural information can be, for example, one or more of the parameters shown in Table 1, Table 2

and Table 3. The various analyses can be performed concurrently or in series. Further, when using two or more indices each of the indices can be weighted equally or differently, or combinations thereof where more than two indices are employed. Additionally, any of these methods can also include analyzing the
5 image for bone mineral density information using any of the methods described herein.

[0031] Any of the methods described herein can further comprise applying one or more correction factors to the data obtained from the image. For example, correction factors can be programmed into a computer unit. The
10 computer unit can be the same one that performs the analysis of the image or can be a different unit. In certain embodiments, the correction factors account for the variation in soft-tissue thickness in individual subjects.

[0032] These and other embodiments of the subject invention will readily occur to those of skill in the art in light of the disclosure herein.

15

BRIEF DESCRIPTION OF THE DRAWINGS

[0033] FIGS. 1A AND B are block diagrams showing the steps for extracting data from an image and then deriving quantitative and/or qualitative data from the image.

[0034] FIGS. 2A-C are diagrams showing an image taken of a region of
20 anatomical interest further illustrating possible locations of regions of interest for analysis.

[0035] FIGS. 3A-J illustrate various abnormalities that might occur including, for example, cartilage defects, bone marrow edema, subchondral sclerosis, osteophytes and cysts.

[0036] FIGS. 4A AND B are block diagrams of the method of FIG. 1A showing that
25 the steps can be repeated.

[0037] FIGS. 5A-E are block diagrams illustrating steps involved in evaluating patterns in an image of a region of interest.

[0038] FIG. 6A-E are block diagrams illustrating steps involved in deriving quantitative and qualitative data from an image in conjunction with administering candidate molecules or drugs for evaluation.

[0039] FIGS. 7A-D are block diagrams illustrating steps involved in comparing derived quantitative and qualitative information to a database or to information obtained at a previous time.

[0040] FIGS. 8A-D are block diagrams illustrating steps involved in comparing converting an image to a pattern of normal and diseased tissue

[0041] FIG. 9 is a diagram showing the use one or more devices in the process of developing a degeneration pattern and using a database for degeneration patterns.

[0042] FIG. 10 depicts regions of interest (ROIs) analyzed in Example 1.

[0043] FIG. 11 depicts results of biomechanical testing of 15 cadaveric hips and femurs.

[0044] FIG. 12A-B, are reproductions of x-ray images depicting an exemplary induced fracture in cadaveric femur resulting from biomechanical testing and load.

[0045] FIG. 13 is a graph depicting correlation of DXA femoral neck bone mineral density (BMD) versus biochemical fracture load as evaluated in 15 fresh cadaveric hip samples.

[0046] FIG. 14A-C are graphs depicting correlation of bone structure versus mechanical fracture load. FIG. 14A depicts correlation of maximum marrow spacing v. fracture load. FIG. 14B depicts correlation of maximum marrow

spacing (log) v. fracture load. FIG. 14C depicts correlation of percentage of trabecular area v. fracture load.

[0047] FIG. 15A-C are graphs depicting correlation of macro-anatomical features versus biomechanical fracture load. FIG. 15A depicts correlation of cortical thickness v. fracture load. FIG. 15B depicts correlation of hip axis length (HAL) v. fracture load. FIG. 15C depicts correlation of cortical thickness (standard deviation) versus fracture load.

[0048] FIG. 16 is a graph depicting multivariate analysis using a combination of bone structural and macro-anatomical parameters and shows the correlation of predicted fracture load to actual fracture load.

DETAILED DESCRIPTION

[0049] The following description is presented to enable any person skilled in the art to make and use the invention. Various modifications to the embodiments described will be readily apparent to those skilled in the art, and the generic principles defined herein can be applied to other embodiments and applications without departing from the spirit and scope of the present invention as defined by the appended claims. Thus, the present invention is not intended to be limited to the embodiments shown, but is to be accorded the widest scope consistent with the principles and features disclosed herein. To the extent necessary to achieve a complete understanding of the invention disclosed, the specification and drawings of all issued patents, patent publications, and patent applications cited in this application are incorporated herein by reference.

[0050] The practice of the present invention employs, unless otherwise indicated, currently conventional methods of imaging and image processing within the skill of the art. Such techniques are explained fully in the literature. See, e.g., WO 02/22014, X-Ray Structure Determination: A Practical Guide, 2nd

Edition, editors Stout and Jensen, 1989, John Wiley & Sons, publisher; Body CT: A Practical Approach, editor Slone, 1999, McGraw-Hill publisher; The Essential Physics of Medical Imaging, editors Bushberg, Seibert, Leidholdt Jr & Boone, 2002, Lippincott, Williams & Wilkins; X-ray Diagnosis: A Physician's Approach, editor Lam, 1998 Springer-Verlag, publisher; Dental Radiology: Understanding the X-Ray Image, editor Laetitia Brocklebank 1997, Oxford University Press publisher; and Digital Image Processing, editor Kenneth R. Castleman, 1996 Prentice Hall, publisher; The Image Processing Handbook, editor John C. Russ, 3rd Edition, 1998, CRC Press; Active Contours: The Application of Techniques from Graphics, Vision, Control Theory and Statistics to Visual Tracking of Shapes in Motion, Editors Andrew Blake, Michael Isard, 1999 Springer Verlag. As will be appreciated by those of skill in the art, as the field of imaging continues to advance methods of imaging currently employed can evolve over time. Thus, any imaging method or technique that is currently employed is appropriate for application of the teachings of this invention as well as techniques that can be developed in the future. A further detailed description of imaging methods is not provided in order to avoid obscuring the invention.

[0051] As shown in **FIG. 1A**, the first step is to locate a part of the body of a subject, for example in a human body, for study **98**. The part of the body located for study is the region of anatomical interest (RAI). In locating a part of the body for study, a determination is made to, for example, take an image or a series of images of the body at a particular location, e.g. hip, dental, spine, etc. Images include, for example, conventional x-ray images, x-ray tomosynthesis, ultrasound (including A-scan, B-scan and C-scan) computed tomography (CT scan), magnetic resonance imaging (MRI), optical coherence tomography, single photon emission tomography (SPECT), and positron emission tomography, or such other imaging tools that a person of skill in the art would find useful in practicing the invention. Once the image is taken, a region of interest (ROI) can be located within the image **100**. Algorithms can be used to automatically place regions of interest in a particular image. See, e.g., Example 1 describing

automatic placement of ROIs in femurs. Image data is extracted from the image **102**. Finally, quantitative and/or qualitative data is extracted from the image data **120**. The quantitative and/or qualitative data extracted from the image includes, for example, the parameters and measurements shown in Table 1, Table 2 or
5 Table 3.

[0052] Each step of locating a part of the body for study **98**, optionally locating a region of interest **100**, obtaining image data **102**, and deriving data **120**, can be repeated one or more times **99,101, 103, 121**, respectively, as desired.

[0053] As shown in FIG. 1B image data can be optionally enhanced **104** by
10 applying image processing techniques, such as noise filtering or diffusion filtering, to facilitate further analysis. Similar to the process shown in FIG. 1A, locating a part of the body for study **98**, optionally locating a region of interest **100**, obtaining image data **102**, enhancing image data **104**, and deriving data **120**, can be repeated one or more times **99,101, 103, 105, 121**,
15 respectively, as desired.

TABLE 1
Representative Parameters Measured with
Quantitative and Qualitative Image Analysis Methods

PARAMETER	MEASUREMENTS
Bone density and microstructural parameters	<ul style="list-style-type: none"> •Calibration phantom equivalent thickness (Average intensity value of the region of interest expressed as thickness of calibration phantom that would produce the equivalent intensity) •Trabecular contrast •Standard deviation of background subtracted ROI •Coefficient of Variation of ROI (Standard deviation / mean) •(Trabecular equivalent thickness / Marrow equivalent thickness) •Fractal dimension •Hough transform •Fourier spectral analysis (Mean transform coefficient absolute value and mean spatial first moment) •Predominant orientation of spatial energy spectrum

PARAMETER	MEASUREMENTS
	<ul style="list-style-type: none"> •Trabecular area (Pixel count of extracted trabeculae) •Trabecular area / Total area •Trabecular perimeter (Count of trabecular pixels with marrow pixels in their neighborhood, proximity or vicinity) •Trabecular distance transform (For each trabecular pixel, calculation of distance to closest marrow pixel) •Marrow distance transform (For each marrow pixel, calculation of distance to closest trabecular pixel) •Trabecular distance transform regional maximal values (mean, min., max, std. Dev). (Describes thickness and thickness variation of trabeculae) •Marrow distance transform regional maximal values (mean, min., max, std. Dev) •Star volume (Mean volume of all the parts of an object which can be seen unobscured from a random point inside the object in all possible directions) •Trabecular Bone Pattern Factor ($TBPf = (P1 - P2) / (A1 - A2)$ where P1 and A1 are the perimeter length and trabecular bone area before dilation and P2 and A2 corresponding values after a single pixel dilation, measure of connectivity) •Connected skeleton count or Trees (T) •Node count (N) •Segment count (S) •Node-to-node segment count (NN) •Node-to-free-end segment count (NF) •Node-to-node segment length (NNL) •Node-to-free-end segment length (NFL) •Free-end-to-free-end segment length (FFL) •Node-to-node total struts length (NN.TSL) •Free-end-to-free-ends total struts length(FF.TSL) •Total struts length (TSL) •FF.TSL/ TSL •NN.TSL/ TSL •Loop count (Lo) •Loop area •Mean distance transform values for each connected skeleton •Mean distance transform values for each segment (Tb.Th) •Mean distance transform values for each node-to-node segment (Tb.Th.NN) •Mean distance transform values for each node-to-free-end segment (Tb.Th.NF) •Orientation (angle) of each segment •Angle between segments •Length-thickness ratios (NNL/Tb.Th.NN) and (NFL/ Tb.Th.NF) •Interconnectivity index (ICI) $ICI = (N * NN) / (T * (NF + 1))$

PARAMETER	MEASUREMENTS
Cartilage and cartilage defect/diseased cartilage parameters	<ul style="list-style-type: none"> •Total cartilage volume •Partial/Focal cartilage volume •Cartilage thickness distribution (thickness map) •Mean cartilage thickness for total region or focal region •Median cartilage thickness for total region or focal region •Maximum cartilage thickness for total region or focal region •Minimum cartilage thickness for total region or focal region •3D cartilage surface information for total region or focal region •Cartilage curvature analysis for total region or focal region •Volume of cartilage defect/diseased cartilage •Depth of cartilage defect/diseased cartilage •Area of cartilage defect/diseased cartilage •2D or 3D location of cartilage defect/diseased cartilage in articular surface •2D or 3D location of cartilage defect/diseased cartilage in relationship to weight-bearing area •Ratio: diameter of cartilage defect or diseased cartilage / thickness of surrounding normal cartilage •Ratio: depth of cartilage defect or diseased cartilage / thickness of surrounding normal cartilage •Ratio: volume of cartilage defect or diseased cartilage / thickness of surrounding normal cartilage •Ratio: surface area of cartilage defect or diseased cartilage / total joint or articular surface area •Ratio: volume of cartilage defect or diseased cartilage / total cartilage volume
Other articular parameters	<ul style="list-style-type: none"> •Presence or absence of bone marrow edema •Volume of bone marrow edema •Volume of bone marrow edema normalized by width, area, size, volume of femoral condyle(s)/tibial plateau/patella – other bones in other joints •Presence or absence of osteophytes •Presence or absence of subchondral cysts •Presence or absence of subchondral sclerosis •Volume of osteophytes •Volume of subchondral cysts •Volume of subchondral sclerosis •Area of bone marrow edema •Area of osteophytes •Area of subchondral cysts •Area of subchondral sclerosis •Depth of bone marrow edema •Depth of osteophytes •Depth of subchondral cysts •Depth of subchondral sclerosis •Volume, area, depth of osteophytes, subchondral cysts, subchondral sclerosis normalized by width, area, size, volume of femoral condyle(s)/tibial plateau/patella – other bones in other joints •Presence or absence of meniscal tear

PARAMETER	MEASUREMENTS
	<ul style="list-style-type: none"> • Presence or absence of cruciate ligament tear • Presence or absence of collateral ligament tear • Volume of menisci • Ratio of volume of normal to torn/damaged or degenerated meniscal tissue • Ratio of surface area of normal to torn/damaged or degenerated meniscal tissue • Ratio of surface area of normal to torn/damaged or degenerated meniscal tissue to total joint or cartilage surface area • Ratio of surface area of torn/damaged or degenerated meniscal tissue to total joint or cartilage surface area • Size ratio of opposing articular surfaces • Meniscal subluxation/dislocation in mm • Index combining different articular parameters which can also include <ul style="list-style-type: none"> ◦ Presence or absence of cruciate or collateral ligament tear ◦ Body mass index, weight, height • 3D surface contour information of subchondral bone • Actual or predicted knee flexion angle during gait cycle (latter based on gait patterns from subjects with matching demographic data retrieved from motion profile database) • Predicted knee rotation during gait cycle • Predicted knee displacement during gait cycle • Predicted load bearing line on cartilage surface during gait cycle and measurement of distance between load bearing line and cartilage defect/diseased cartilage • Predicted load bearing area on cartilage surface during gait cycle and measurement of distance between load bearing area and cartilage defect/diseased cartilage • Predicted load bearing line on cartilage surface during standing or different degrees of knee flexion and extension and measurement of distance between load bearing line and cartilage defect/diseased cartilage • Predicted load bearing area on cartilage surface during standing or different degrees of knee flexion and extension and measurement of distance between load bearing area and cartilage defect/diseased cartilage • Ratio of load bearing area to area of cartilage defect/diseased cartilage • Percentage of load bearing area affected by cartilage disease • Location of cartilage defect within load bearing area • Load applied to cartilage defect, area of diseased cartilage • Load applied to cartilage adjacent to cartilage defect, area of diseased cartilage

[0054] As will be appreciated by those of skill in the art, the parameters and measurements shown in Table 1 are provided for illustration purposes. It will be apparent that the terms micro-structural parameters, micro-architecture, micro-

anatomic structure, micro-structural and trabecular architecture may be used interchangeably. In addition, other parameters and measurements, ratios, derived values or indices can be used to extract quantitative and/or qualitative information about the ROI without departing from the scope of the invention.

- 5 Additionally, where multiple ROI or multiple derivatives of data are used, the parameter measured can be the same parameter or a different parameter without departing from the scope of the invention. Additionally, data from different ROIs can be combined or compared as desired.

[0055] Additional measurements can be performed that are selected based on
10 the anatomical structure to be studied as described below.

[0056] Once the data is extracted from the image it can be manipulated to assess the severity of the disease and to determine disease staging (e.g., mild, moderate, severe or a numerical value or index). The information can also be used to monitor progression of the disease and/or the efficacy of any
15 interventional steps that have been taken. Finally, the information can be used to predict the progression of the disease or to randomize patient groups in clinical trials.

[0057] FIG. 2A illustrates an image **200** taken of an RAI, shown as **202**. As shown in FIG. 2A, a single region of interest (ROI) **210** has been identified within
20 the image. The ROI **210** can take up the entire image **200**, or nearly the entire image. As shown in FIG. 2B more than one ROI can be identified in an image. In this example, a first ROI **220** is depicted in one region of the image **200**, and a second ROI **222** is depicted within the image. In this instance, neither of these ROI overlap or abut. As will be appreciated by a person of skill in the art, the
25 number of ROI identified in an image **200** is not limited to the two depicted. Turning now to FIG. 2c another embodiment showing two ROI for illustration purposes is shown. In this instance, the first ROI **230** and the second ROI **232**, are partially overlapping. As will be appreciated by those of skill in the art, where

multiple ROI are used any or all of the ROI can be organized such that it does not overlap, it abuts without overlapping, it overlaps partially, it overlaps completely (for example where a first ROI is located completely within a second identified ROI), and combinations thereof. Further the number of ROI per image
5 **200** can range from one (ROI_1) to n (ROI_n) where n is the number of ROI to be analyzed.

[0058] Bone density, microarchitecture, macro-anatomic and/or biomechanical (e.g. derived using finite element modeling) analyses can be applied within a region of predefined size and shape and position. This region of interest can
10 also be referred to as a “window.” Processing can be applied repeatedly within the window at different positions of the image. For example, a field of sampling points can be generated and the analysis performed at these points. The results of the analyses for each parameter can be stored in a matrix space, e.g., where its position corresponds to the position of the sampling point where the analysis
15 occurred, thereby forming a map of the spatial distribution of the parameter (a parameter map). The sampling field can have regular intervals or irregular intervals with varying density across the image. The window can have variable size and shape, for example to account for different patient size or anatomy.

[0059] The amount of overlap between the windows can be determined, for
20 example, using the interval or density of the sampling points (and resolution of the parameter maps). Thus, the density of sampling points is set higher in regions where higher resolution is desired and set lower where moderate resolution is sufficient, in order to improve processing efficiency. The size and shape of the window would determine the local specificity of the parameter.
25 Window size is preferably set such that it encloses most of the structure being measured. Oversized windows are generally avoided to help ensure that local specificity is not lost.

[0060] The shape of the window can be varied to have the same orientation and/or geometry of the local structure being measured to minimize the amount of structure clipping and to maximize local specificity. Thus, both 2D and/or 3D windows can be used, as well as combinations thereof, depending on the nature
5 of the image and data to be acquired.

[0061] In another embodiment, bone density, microarchitecture, macro-anatomic and/or biomechanical (e.g. derived using finite element modeling) analyses can be applied within a region of predefined size and shape and position. The region is generally selected to include most, or all, of the anatomic
10 region under investigation and, preferably, the parameters can be assessed on a pixel-by-pixel basis (e.g., in the case of 2D or 3D images) or a voxel-by-voxel basis in the case of cross-sectional or volumetric images (e.g., 3D images obtained using MR and/or CT). Alternatively, the analysis can be applied to clusters of pixels or voxels wherein the size of the clusters is typically selected to
15 represent a compromise between spatial resolution and processing speed. Each type of analysis can yield a parameter map.

[0062] Parameter maps can be based on measurement of one or more parameters in the image or window; however, parameter maps can also be derived using statistical methods. In one embodiment, such statistical
20 comparisons can include comparison of data to a reference population, e.g. using a z-score or a T-score. Thus, parameter maps can include a display of z-scores or T-scores.

[0063] Additional measurements relating to the site to be measured can also be taken. For example, measurements can be directed to dental, spine, hip, knee
25 or bone cores. Examples of suitable site specific measurements are shown in Table 2.

TABLE 2
Site specific measurement of bone parameters

<u>Parameters specific to hip images</u>	<ul style="list-style-type: none"> •All microarchitecture parameters on structures parallel to stress lines •All microarchitecture parameters on structures perpendicular to stress lines <p>Geometry</p> <ul style="list-style-type: none"> •Shaft angle •Neck angle •Average and minimum diameter of femur neck •Hip axis length •CCD (caput-collum-diaphysis) angle •Width of trochanteric region •Largest cross-section of femur head •Standard deviation of cortical bone thickness within ROI •Minimum, maximum, mean and median thickness of cortical bone within ROI •Hip joint space width
<u>Parameters specific to spine images</u>	<ul style="list-style-type: none"> •All microarchitecture parameters on vertical structures •All microarchitecture parameters on horizontal structures <p>Geometry</p> <ul style="list-style-type: none"> •Superior endplate cortical thickness (anterior, center, posterior) •Inferior endplate cortical thickness (anterior, center, posterior) •Anterior vertebral wall cortical thickness (superior, center, inferior) •Posterior vertebral wall cortical thickness (superior, center, inferior) •Superior aspect of pedicle cortical thickness •inferior aspect of pedicle cortical thickness •Vertebral height (anterior, center, posterior) •Vertebral diameter (superior, center, inferior), •Pedicle thickness (supero-inferior direction). •Maximum vertebral height •Minimum vertebral height •Average vertebral height •Anterior vertebral height •Medial vertebral height •Posterior vertebral height •Maximum inter-vertebral height •Minimum inter-vertebral height •Average inter-vertebral height
<u>Parameters specific to knee images</u>	<ul style="list-style-type: none"> •Average medial joint space width •Minimum medial joint space width •Maximum medial joint space width •Average lateral joint space width •Minimum lateral joint space width •Maximum lateral joint space width

[0064] As will be appreciated by those of skill in the art, measurement and image processing techniques are adaptable to be applicable to both microarchitecture and macro-anatomical structures. Examples of these measurements are shown in Table 3.

5

TABLE 3
Measurements applicable on Microarchitecture and Macro-anatomical Structures

<u>Average density measurement</u>	<ul style="list-style-type: none"> •Calibrated density of ROI
<u>Measurements on micro-anatomical structures of dental, spine, hip, knee or bone cores images</u>	<p>The following parameters are derived from the extracted structures:</p> <ul style="list-style-type: none"> •Calibrated density of extracted structures •Calibrated density of background •Average intensity of extracted structures •Average intensity of background (area other than extracted structures) •Structural contrast (average intensity of extracted structures / average intensity of background) •Calibrated structural contrast (calibrated density extracted structures / calibrated density of background) •Total area of extracted structures •Total area of ROI •Area of extracted structures normalized by total area of ROI •Boundary lengths (perimeter) of extracted normalized by total area of ROI •Number of structures normalized by area of ROI •Trabecular bone pattern factor; measures concavity and convexity of structures •Star volume of extracted structures •Star volume of background •Number of loops normalized by area of ROI
<u>Measurements on Distance transform of extracted structures</u>	<p>The following statistics are measured from the distance transform regional maximum values:</p> <ul style="list-style-type: none"> •Average regional maximum thickness •Standard deviation of regional maximum thickness •Largest value of regional maximum thickness •Median of regional maximum thickness
<u>Measurements on skeleton of extracted structures</u>	<ul style="list-style-type: none"> •Average length of networks (units of connected segments) •Maximum length of networks •Average thickness of structure units (average distance transform values along skeleton) •Maximum thickness of structure units (maximum distance transform values along skeleton) •Number of nodes normalized by ROI area •Number of segments normalized by ROI area •Number of free-end segments normalized by ROI area

	<ul style="list-style-type: none"> •Number of inner (node-to-node) segments normalized ROI area •Average segment lengths •Average free-end segment lengths •Average inner segment lengths •Average orientation angle of segments •Average orientation angle of inner segments •Segment tortuosity; a measure of straightness •Segment solidity; another measure of straightness •Average thickness of segments (average distance transform values along skeleton segments) •Average thickness of free-end segments •Average thickness of inner segments •Ratio of inner segment lengths to inner segment thickness •Ratio of free-end segment lengths to free-end segment thickness •Interconnectivity index; a function of number of inner segments, free-end segments and number of networks.
Directional skeleton segment measurements	All measurement of skeleton segments can be constrained by one or more desired orientation by measuring only skeleton segments within ranges of angle.
Watershed segmentation	<p>Watershed segmentation is applied to gray level images. Statistics of watershed segments are:</p> <ul style="list-style-type: none"> •Total area of segments •Number of segments normalized by total area of segments •Average area of segments •Standard deviation of segment area •Smallest segment area •Largest segment area

[0065] As noted above, analysis can also include one or more additional techniques include, for example, Hough transform, mean pixel intensity analysis, variance of pixel intensity analysis, soft tissue analysis and the like. See, e.g.,
5 co-owned International Application WO 02/30283.

[0066] Calibrated density typically refers to the measurement of intensity values of features in images converted to its actual material density or expressed as the density of a reference material whose density is known. The reference material can be metal, polymer, plastics, bone, cartilage, etc., and can be part of
10 the object being imaged or a calibration phantom placed in the imaging field of view during image acquisition.

[0067] Extracted structures typically refer to simplified or amplified representations of features derived from images. An example would be binary images of trabecular patterns generated by background subtraction and thresholding. Another example would be binary images of cortical bone generated by applying an edge filter and thresholding. The binary images can be superimposed on gray level images to generate gray level patterns of structure of interest.

[0068] Distance transform typically refers to an operation applied on binary images where maps representing distances of each 0 pixel to the nearest 1 pixel are generated. Distances can be calculated by the Euclidian magnitude, city-block distance, La Place distance or chessboard distance.

[0069] Distance transform of extracted structures typically refer to distance transform operation applied to the binary images of extracted structures, such as those discussed above with respect to calibrated density.

[0070] Skeleton of extracted structures typically refer to a binary image of 1 pixel wide patterns, representing the centerline of extracted structures. It is generated by applying a skeletonization or medial transform operation, by mathematical morphology or other methods, on an image of extracted structures.

[0071] Skeleton segments typically are derived from skeleton of extracted structures by performing pixel neighborhood analysis on each skeleton pixel. This analysis classifies each skeleton pixel as a node pixel or a skeleton segment pixel. A node pixel has more than 2 pixels in its 8-neighborhood. A skeleton segment is a chain of skeleton segment pixels continuously 8-connected. Two skeleton segments are separated by at least one node pixel.

[0072] Watershed segmentation as it is commonly known to a person of skill in the art, typically is applied to gray level images to characterize gray level continuity of a structure of interest. The statistics of dimensions of segments

generated by the process are, for example, those listed in Table 3 above. As will be appreciated by those of skill in the art, however, other processes can be used without departing from the scope of the invention.

5 [0073] Turning now to FIG. 3A, a cross-section of a cartilage defect is shown **300**. The cross-hatched zone **302** corresponds to an area where there is cartilage loss. FIG. 3B is a top view of the cartilage defect shown in FIG. 3A.

[0074] FIG. 3C illustrates the depth of a cartilage defect **310** in a first cross-section dimension with a dashed line illustrating a projected location of the original cartilage surface **312**. By comparing these two values a ratio of cartilage defect depth to cartilage defect width can be calculated.

[0075] FIG. 3D illustrated the depth of the cartilage **320** along with the width of the cartilage defect **322**. These two values can be compared to determine a ratio of cartilage depth to cartilage defect width.

15 [0076] FIG. 3E shows the depth of the cartilage defect **310** along with the depth of the cartilage **320**. A dashed line is provided illustrating a projected location for the original cartilage surface **312**. Similar to the measurements made above, ratios between the various measurements can be calculated.

20 [0077] Turning now to FIG. 3F, an area of bone marrow edema is shown on the femur **330** and the tibia **332**. The shaded area of edema can be measured on a T2-weighted MRI scan. Alternatively, the area can be measured on one or more slices. These measurements can then be extended along the entire joint using multiple slices or a 3D acquisition. From these measurements volume can be determined or derived.

25 [0078] FIG. 3G shows an area of subchondral sclerosis in the acetabulum **340** and the femur **342**. The sclerosis can be measured on, for example, a T1 or T2-weighted MRI scan or on a CT scan. The area can be measured on one or more slices. Thereafter the measurement can be extended along the entire joint using

multiple slices or a 3D acquisition. From these values a volume can be derived of the subchondral sclerosis. For purposes of illustration, a single sclerosis has been shown on each surface. However, a person of skill in the art will appreciate that more than one sclerosis can occur on a single joint surface.

5 **[0079]** **FIG.3H** shows osteophytes on the femur **350** and the tibia **352**. The osteophytes are shown as cross-hatched areas. Similar to the sclerosis shown in **FIG. 3G**, the osteophytes can be measured on, for example, a T1 or T2-weighted MRI scan or on a CT scan. The area can be measured on one or more slices. Thereafter the measurement can be extended along the entire joint using
10 multiple slices or a 3D acquisition. From these values a volume can be derived of the osteophytes. Additionally, a single osteophyte **354** or osteophyte groups **356** can be included in any measurement. Persons of skill in the art will appreciate that groups can be taken from a single joint surface or from opposing joint surfaces, as shown, without departing from the scope of the invention.

15 **[0080]** Turning now to **FIG. 3I** an area of subchondral cysts **360, 362, 364** is shown. Similar to the sclerosis shown in **FIG. 3G**, the cysts can be measured on, for example, a T1 or T2-weighted MRI scan or on a CT scan. The area can be measured on one or more slices. Thereafter the measurement can be extended along the entire joint using multiple slices or a 3D acquisition. From these values
20 a volume can be derived of the cysts. Additionally, single cysts **366** or groups of cysts **366'** can be included in any measurement. Persons of skill in the art will appreciate that groups can be taken from a single joint surface, as shown, or from opposing joint surfaces without departing from the scope of the invention.

25 **[0081]** **FIG. 3J** illustrates an area of torn meniscal tissue (cross-hatched) **372, 374** as seen from the top **370** and in cross-section **371**. Again, similar to the sclerosis shown in **FIG. 3G**, the torn meniscal tissue can be measured on, for example, a T1 or T2-weighted MRI scan or on a CT scan. The area can be measured on one or more slices. Thereafter the measurement can be extended

along the entire joint using multiple slices or a 3D acquisition. From these values a volume can be derived of the tear. Ratios such as surface or volume of torn to normal meniscal tissue can be derived as well as ratios of surface of torn meniscus to surface of opposing articulating surface.

5 **[0082]** As shown in **FIG. 4A**, the process of optionally locating a ROI **100**,
extracting image data from the ROI **102**, and deriving quantitative and/or
qualitative image data from the extracted image data **120**, can be repeated **122**.
Alternatively, or in addition, the process of locating a ROI **100**, can be repeated
124. A person of skill in the art will appreciate that these steps can be repeated
10 one or more times in any appropriate sequence, as desired, to obtain a sufficient
amount of quantitative and/or qualitative data on the ROI or to separately extract
or evaluate parameters. Further, the ROI used can be the same ROI as used in
the first process or a newly identified ROI in the image. Additionally, as with
FIG. 1A the steps of locating a region of interest **100**, obtaining image data **102**,
15 and deriving quantitative and/or qualitative image data can be repeated one or
more times, as desired, **101**, **103**, **121**, respectively. Although not depicted here,
as discussed above with respect to **FIG. 1A**, the additional step of locating a part
of the body for study **98** can be performed prior to locating a region of interest
100 without departing from the invention. Additionally that step can be repeated
20 **99**.

[0083] **FIG. 4B** illustrates the process shown in **FIG. 4A** with the additional step
enhancing image data **104**. Additionally, the step of enhancing image data **104**
can be repeated one or more times **105**, as desired. The process of enhancing
image data **104** can be repeated **126** one or more times as desired.

25 **[0084]** Turning now to **FIG. 5A**, a process is shown whereby a region of
interest is optionally located **100**. Although not depicted here, as discussed
above with respect to **FIG. 1A**, the step of locating a part of the body for study **98**
can be performed prior to locating a region of interest **100** without departing from

the invention. Additionally that step can be repeated **99**. Once the region of interest is located **100**, and image data is extracted from the ROI **102**, the extracted image data can then be converted to a 2D pattern **130**, a 3D pattern **132** or a 4D pattern **133**, for example including velocity or time, to facilitate data analyses. Following conversion to 2D **130**, 3D **132** or 4D pattern **133** the images are evaluated for patterns **140**. Additionally images can be converted from 2D to 3D **131**, or from 3D to 4D **131'**, if desired. Although not illustrated to avoid obscuring the figure, persons of skill in the art will appreciate that similar conversions can occur between 2D and 4D in this process or any process illustrated in this invention.

[0085] As will be appreciated by those of skill in the art, the conversion step is optional and the process can proceed directly from extracting image data from the ROI **102** to evaluating the data pattern **140** directly **134**. Evaluating the data for patterns, includes, for example, performing the measurements described in Table 1, Table 2 or Table 3, above.

[0086] Additionally, the steps of locating the region of interest **100**, obtaining image data **102**, and evaluating patterns **141** can be performed once or a plurality of times, **101**, **103**, **141**, respectively at any stage of the process. As will be appreciated by those of skill in the art, the steps can be repeated. For example, following an evaluation of patterns **140**, additional image data can be obtained **135**, or another region of interest can be located **137**. These steps can be repeated as often as desired, in any combination desirable to achieve the data analysis desired.

[0087] FIG. 5B illustrates an alternative process to that shown in FIG. 5A which ~~5A~~ THAT includes the step of enhancing image data **104** prior to converting an image or image data to a 2D **130**, 3D **132**, or 4D **133** pattern. The process of enhancing image data **104**, can be repeated **105** if desired. FIG. 5c illustrates an alternative embodiment to the process shown in FIG. 5B. In this process, the step

of enhancing image data **104** occurs after converting an image or image data to a 2D **130**, 3D **132**, or 4D **133** pattern. Again, the process of enhancing image data **104**, can be repeated **105** if desired.

[0088] FIG. 5D illustrates an alternative process to that shown in FIG. 5A. After
5 locating a part of the body for study **98** and imaging, the image is then converted to a 2D pattern **130**, 3D pattern **132** or 4D pattern **133**. The region of interest **100** is optionally located within the image after conversion to a 2D, 3D or 4D image and data is then extracted **102**. Patterns are then evaluated in the extracted image data **140**. As with the process of FIG. 5A, the conversion step is optional.
10 Further, if desired, images can be converted between 2D, 3D **131** and 4D **131'** if desired.

[0089] Similar to FIG. 5A, some or all the processes can be repeated one or more times as desired. For example, locating a part of the body for study **98**, locating a region of interest **100**, obtaining image data **102**, and evaluating
15 patterns **140**, can be repeated one or more times if desired, **99**, **101**, **103**, **141**, respectively. Again steps can be repeated. For example, following an evaluation of patterns **140**, additional image data can be obtained **135**, or another region of interest can be located **137** and/or another portion of the body can be located for study **139**. These steps can be repeated as often as desired, in any combination
20 desirable to achieve the data analysis desired.

[0090] FIG. 5E illustrates an alternative process to that shown in FIG. 5D. In this process image data can be enhanced **104**. The step of enhancing image data can occur prior to conversion **143**, prior to locating a region of interest **145**, prior to obtaining image data **102**, or prior to evaluating patterns **149**.

25 **[0091]** Similar to FIG. 5A, some or all the processes can be repeated one or more times as desired, including the process of enhancing image data **104**, which is shown as **105**.

[0092] The method also comprises obtaining an image of a bone or a joint, optionally converting the image to a two-dimensional or three-dimensional or four-dimensional pattern, and evaluating the amount or the degree of normal, diseased or abnormal tissue or the degree of degeneration in a region or a volume of interest using one or more of the parameters specified in Table 1, Table 2 and/or Table 3. By performing this method at an initial time T_1 , information can be derived that is useful for diagnosing one or more conditions or for staging, or determining, the severity of a condition. This information can also be useful for determining the prognosis of a patient, for example with osteoporosis or arthritis. By performing this method at an initial time T_1 , and a later time T_2 , the change, for example in a region or volume of interest, can be determined which then facilitates the evaluation of appropriate steps to take for treatment. Moreover, if the subject is already receiving therapy or if therapy is initiated after time T_1 , it is possible to monitor the efficacy of treatment. By performing the method at subsequent times, T_2 - T_n , additional data can be acquired that facilitate predicting the progression of the disease as well as the efficacy of any interventional steps that have been taken. As will be appreciated by those of skill in the art, subsequent measurements can be taken at regular time intervals or irregular time intervals, or combinations thereof. For example, it can be desirable to perform the analysis at T_1 with an initial follow-up, T_2 , measurement taken one month later. The pattern of one month follow-up measurements could be performed for a year (12 one-month intervals) with subsequent follow-ups performed at 6 month intervals and then 12 month intervals. Alternatively, as an example, three initial measurements could be at one month, followed by a single six month follow up which is then followed again by one or more one month follow-ups prior to commencing 12 month follow ups. The combinations of regular and irregular intervals are endless, and are not discussed further to avoid obscuring the invention.

[0093] Moreover, one or more of the parameters listed in Tables 1, 2 and 3 can be measured. The measurements can be analyzed separately or the data

can be combined, for example using statistical methods such as linear regression modeling or correlation. Actual and predicted measurements can be compared and correlated. See, *also*, Example 1.

[0094] The method for assessing the condition of a bone or joint in a subject
5 can be fully automated such that the measurements of one or more of the
parameters specified in Table 1, Table 2 or Table 3 are done automatically
without intervention. The automatic assessment then can include the steps of
diagnosis, staging, prognostication or monitoring the disease or diseases, or to
monitor therapy. As will be appreciated by those of skill in the art, the fully
10 automated measurement is, for example, possible with image processing
techniques such as segmentation and registration. This process can include, for
example, seed growing, thresholding, atlas and model based segmentation
methods, live wire approaches, active and/or deformable contour approaches,
contour tracking, texture based segmentation methods, rigid and non-rigid
15 surface or volume registration, for example based on mutual information or other
similarity measures. One skilled in the art will readily recognize other techniques
and methods for fully automated assessment of the parameters and
measurements specified in Table 1, Table 2 and Table 3.

[0095] Alternatively, the method of assessing the condition of a bone or joint in
20 a subject can be semi-automated such that the measurements of one or more of
the parameters, such as those specified in Table 1, are performed semi-
automatically, i.e., with intervention. The semi-automatic assessment then allows
for human interaction and, for example, quality control, and utilizing the
measurement of said parameter(s) to diagnose, stage, prognosticate or monitor a
25 disease or to monitor a therapy. The semi-automated measurement is, for
example, possible with image processing techniques such as segmentation and
registration. This can include seed growing, thresholding, atlas and model based
segmentation methods, live wire approaches, active and/or deformable contour
approaches, contour tracking, texture based segmentation methods, rigid and

non-rigid surface or volume registration, for example base on mutual information or other similarity measures. One skilled in the art will readily recognize other techniques and methods for semi-automated assessment of the parameters specified in Table 1, Table 2 or Table 3.

5 **[0096]** Turning now to **FIG. 6A**, a process is shown whereby the user locates a ROI **100**, extracts image data from the ROI **102**, and then derives quantitative and/or qualitative image data from the extracted image data **120**, as shown above with respect to **FIG. 1**. Following the step of deriving quantitative and/or qualitative image data, a candidate agent is administered to the patient **150**. The
10 candidate agent can be any agent the effects of which are to be studied. Agents can include any substance administered or ingested by a subject, for example, molecules, pharmaceuticals, biopharmaceuticals, agropharmaceuticals, or combinations thereof, including cocktails, that are thought to affect the quantitative and/or qualitative parameters that can be measured in a region of
15 interest. These agents are not limited to those intended to treat disease that affects the musculoskeletal system but this invention is intended to embrace any and all agents regardless of the intended treatment site. Thus, appropriate agents are any agents whereby an effect can be detected via imaging. The steps of locating a region of interest **100**, obtaining image data **102**, obtaining
20 quantitative and/or qualitative data from image data **120**, and administering a candidate agent **150**, can be repeated one or more times as desired, **101, 103, 121, 151**, respectively.

[0097] **FIG. 6B** shows the additional step of enhancing image data **104**, which can also be optionally repeated **105** as often as desired.

25 **[0098]** As shown in **FIG. 6c** these steps can be repeated one or more times **152** to determine the effect of the candidate agent. As will be appreciated by those of skill in the art, the step of repeating can occur at the stage of locating a region of interest **152** as shown in **FIG. 6B** or it can occur at the stage obtaining

image data **153** or obtaining quantitative and/or qualitative data from image data **154** as shown in FIG. 6D.

[0099] FIG. 6E shows the additional step of enhancing image data **104**, which can optionally be repeated **105**, as desired.

5 [0100] As previously described, some or all the processes shown in FIGS. 6A-E can be repeated one or more times as desired. For example, locating a region of interest **100**, obtaining image data **102**, enhancing image data **104**, obtaining quantitative and/or qualitative data **120**, evaluating patterns **140**, and administering candidate agent **150** can be repeated one or more times if desired,
10 **101, 103, 105, 121, 141, 151** respectively.

[0101] In the scenario described in relation to FIGS. 6, an image is taken prior to administering the candidate agent. However, as will be appreciated by those of skill in the art, it is not always possible to have an image prior to administering the candidate agent. In those situations, progress is determined over time by
15 evaluating the change in parameters from extracted image to extracted image.

[0102] Turning now to FIG. 7A, the process is shown whereby the candidate agent is administered first **150**. Thereafter a region of interest is located in an image taken **100** and image data is extracted **102**. Once the image data is extracted, quantitative and/or qualitative data is extracted from the image data
20 **120**. In this scenario, because the candidate agent is administered first, the derived quantitative and/or qualitative data derived is compared to a database **160** or a subset of the database, which database that, includes data for subjects having similar tracked parameters. As shown in FIG. 7B following the step of obtaining image data, the image data can be enhanced **104**. This process
25 can optionally be repeated **105**, as desired.

[0103] Alternatively, as shown in FIG. 7C the derived quantitative and/or qualitative information can be compared to an image taken at T1 **162**, or any

other time, if such image is available. As shown in FIG. 7D the step of enhancing image data **104** can follow the step of obtaining image data **102**. Again, the process can be repeated **105**, as desired.

[0104] As previously described, some or all the processes illustrated in

5 **FIGS. 7A-D** can be repeated one or more times as desired. For example, locating a region of interest **100**, obtaining image data **102**, enhancing image data **104**, obtaining quantitative and/or qualitative data **120**, administering candidate agent **150**, comparing quantitative and/or qualitative information to a database **160**, comparing quantitative and/or qualitative information to an image taken at a prior
10 time, such as T_1 , **162**, monitoring therapy **170**, monitoring disease progress **172**, predicting disease course **174** can be repeated one or more times if desired, **101**, **103**, **105**, **121**, **151**, **161**, **163**, **171**, **173**, **175** respectively. Each of these steps can be repeated in one or more loops as shown in **FIG. 7B**, **176**, **177**, **178**, **179**, **180**, as desired or appropriate to enhance data collection.

15 **[0105]** Turning now to **FIG. 8A**, following the step of extracting image data from the ROI **102**, the image can be transmitted **180**. Transmission can be to another computer in the network or via the World Wide Web to another network.

Following the step of transmitting the image **180**, the image is converted to a pattern of normal and diseased tissue **190**. Normal tissue includes the

20 undamaged tissue located in the body part selected for study. Diseased tissue includes damaged tissue located in the body part selected for study. Diseased tissue can also include, or refer to, a lack of normal tissue in the body part selected for study. For example, damaged or missing cartilage would be considered diseased tissue. Once the image is converted, it is analyzed **200**.

25 **FIG. 8B** illustrates the process shown in **FIG. 8A** with the additional step of enhancing image data **104**. As will be appreciated by those of skill in the art, this process can be repeated **105** as desired.

[0106] As shown in **FIG. 8c**, the step of transmitting the image **180** illustrated in **FIG. 8A** is optional and need not be practiced under the invention. As will be appreciated by those of skill in the art, the image can also be analyzed prior to converting the image to a pattern of normal and diseased. **FIG. 8D** illustrates the process shown in **FIG. 8c** with the additional step of enhancing image data **104** that is optionally repeated **105**, as desired.

[0107] As previously described, some or all the processes in **FIGS. 8A-D** can be repeated one or more times as desired. For example, locating a region of interest **100**, obtaining image data **102**, enhancing image data **104**, transmitting an image **180**, converting the image to a pattern of normal and diseased **190**, analyzing the converted image **200**, can be repeated one or more times if desired, **101, 103, 105, 181, 191, 201** respectively.

[0108] **FIG. 9** shows two devices **900, 920** that are connected. Either the first or second device can develop a degeneration pattern from an image of a region of interest **905**. Similarly, either device can house a database for generating additional patterns or measurements **915**. The first and second devices can communicate with each other in the process of analyzing an image, developing a degeneration pattern from a region of interest in the image, and creating a dataset of patterns or measurements or comparing the degeneration pattern to a database of patterns or measurements. However, all processes can be performed on one or more devices, as desired or necessary.

[0109] In this method the electronically generated, or digitized image or portions of the image can be electronically transferred from a transferring device to a receiving device located distant from the transferring device; receiving the transferred image at the distant location; converting the transferred image to a pattern of normal or diseased or abnormal tissue using one or more of the parameters specified in Table 1, Table 2 or Table 3; and optionally transmitting the pattern to a site for analysis. As will be appreciated by those of skill in the art,

the transferring device and receiving device can be located within the same room or the same building. The devices can be on a peer-to-peer network, or an intranet. Alternatively, the devices can be separated by large distances and the information can be transferred by any suitable means of data transfer, including
5 the World Wide Web and ftp protocols.

[0110] Alternatively, the method can comprise electronically transferring an electronically-generated image or portions of an image of a bone or a joint from a transferring device to a receiving device located distant from the transferring device; receiving the transferred image at the distant location; converting the
10 transferred image to a degeneration pattern or a pattern of normal or diseased or abnormal tissue using one or more of the parameters specified in Table 1, Table 2 or Table 3; and optionally transmitting the degeneration pattern or the pattern of normal or diseased or abnormal tissue to a site for analysis.

[0111] Thus, the invention described herein includes methods and systems
15 for prognosis of musculoskeletal disease, for example prognosis of fracture risk and the like. (See, *also*, Example 1). FIG. 10 is a schematic depiction of an image of a femur showing various ROIs that were analyzed to predict fracture risk based on assessment of one or more parameters shown in Tables 1, 2 and 3.

[0112] In order to make more accurate prognoses, it may be desirable in
20 certain instances to compare data obtained from a subject to a reference database. For example, when predicting fracture risk, it may be useful to compile data of actual (known) fracture load in a variety of samples and store the results based on clinical risk factors such as age, sex and weight (or other
25 characteristics) of the subject from which the sample is obtained. The images of these samples are analyzed to obtain parameters shown in Tables 1, 2 and 3. A fracture risk model correlated with fracture load may be developed using univariate, bivariate and/or multivariate statistical analysis of these parameters

and is stored in this database. A fracture risk model may include information that is used to estimate fracture risk from parameters shown in Tables 1, 2 and 3. An example of a fracture risk model is the coefficients of a multivariate linear model derived from multivariate linear regression of these parameters (Tables 1,2,3, age, sex, weight, etc.) with fracture load. A person skilled in the art will appreciate that fracture risk models can be derived using other methods such as artificial neural networks and be represented by other forms such as the coefficients of artificial neural networks. Patient fracture risk can then be determined from measurements obtain from bone images by referencing to this database.

[0113] Methods of determining actual fracture load are known to those in the field. FIG. 11 is a schematic depiction of biomechanical testing of an intact femur. As shown, cross-sectional images may be taken throughout testing to determine at what load force a fracture occurs. FIG. 12B is a reproduction of an x-ray image depicting an example of an induced fracture in a fresh cadaveric femur.

[0114] The analysis techniques described herein can then be applied to a subject and the risk of fracture (or other disease) predicted using one or more of the parameters described herein. As shown in FIGS. 13 to 16, the prognostication methods described herein are as (or more) accurate than known techniques in predicting fracture risk. FIG. 13 is a graph depicting linear regression analysis of DXA bone mineral density correlated to fracture load. Correlations of individual parameters to fracture load are comparable to DXA (FIG. 14 and 15). However, when multiple structural parameters are combined, the prediction of load at which fracture will occur is more accurate. (FIG. 16). Thus, the analyses of images as described herein can be used to accurately predict musculoskeletal disease such as fracture risk.

[0115] Another aspect of the invention is a kit for aiding in assessing the condition of a bone or a joint of a subject, which kit comprises a software

program, which when installed and executed on a computer reads a degeneration pattern or a pattern of normal or diseased or abnormal tissue derived using one or more of the parameters specified in Table 1, Table 2 or Table 3 presented in a standard graphics format and produces a computer
5 readout. The kit can further include a database of measurements for use in calibrating or diagnosing the subject. One or more databases can be provided to enable the user to compare the results achieved for a specific subject against, for example, a wide variety of subjects, or a small subset of subjects having characteristics similar to the subject being studied.

- 10 **[0116]** A system is provided that includes (a) a device for electronically transferring a degeneration pattern or a pattern of normal, diseased or abnormal tissue for the bone or the joint to a receiving device located distant from the transferring device; (b) a device for receiving said pattern at the remote location; (c) a database accessible at the remote location for generating additional
15 patterns or measurements for the bone or the joint of the human wherein the database includes a collection of subject patterns or data, for example of human bones or joints, which patterns or data are organized and can be accessed by reference to characteristics such as type of joint, gender, age, height, weight, bone size, type of movement, and distance of movement; (d) optionally a device
20 for transmitting the correlated pattern back to the source of the degeneration pattern or pattern of normal, diseased or abnormal tissue.

- [0117]** Thus, the methods and systems described herein make use of collections of data sets of measurement values, for example measurements of bone structure and/or bone mineral density from images (e.g., x-ray images).
25 Records can be formulated in spreadsheet-like format, for example including data attributes such as date of image (x-ray), patient age, sex, weight, current medications, geographic location, etc. The database formulations can further comprise the calculation of derived or calculated data points from one or more acquired data points, typically using the parameters listed in Tables 1, 2 and 3 or

combinations thereof. A variety of derived data points can be useful in providing information about individuals or groups during subsequent database manipulation, and are therefore typically included during database formulation.

Derived data points include, but are not limited to the following: (1) maximum

5 value, e.g. bone mineral density, determined for a selected region of bone or joint or in multiple samples from the same or different subjects; (2) minimum value, e.g. bone mineral density, determined for a selected region of bone or joint or in multiple samples from the same or different subjects; (3) mean value, e.g. bone mineral density, determined for a selected region of bone or joint or in multiple

10 samples from the same or different subjects; (4) the number of measurements that are abnormally high or low, determined by comparing a given measurement data point with a selected value; and the like. Other derived data points include, but are not limited to the following: (1) maximum value of a selected bone structure parameter, determined for a selected region of bone or in multiple

15 samples from the same or different subjects; (2) minimum value of a selected bone structure parameter, determined for a selected region of bone or in multiple samples from the same or different subjects; (3) mean value of a selected bone structure parameter, determined for a selected region of bone or in multiple samples from the same or different subjects; (4) the number of bone structure

20 measurements that are abnormally high or low, determined by comparing a given measurement data point with a selected value; and the like. Other derived data points will be apparent to persons of ordinary skill in the art in light of the teachings of the present specification. The amount of available data and data derived from (or arrived at through analysis of) the original data provides an

25 unprecedented amount of information that is very relevant to management of bone-related diseases such as osteoporosis. For example, by examining subjects over time, the efficacy of medications can be assessed.

[0118] Measurements and derived data points are collected and calculated, respectively, and can be associated with one or more data attributes to form a

30 database. The amount of available data and data derived from (or arrived at

through analysis of) the original data provide provides an unprecedented amount of information that is very relevant to management of musculoskeletal-related diseases such as osteoporosis or arthritis. For example, by examining subjects over time, the efficacy of medications can be assessed.

- 5 **[0119]** Data attributes can be automatically input with the electronic image and can include, for example, chronological information (e.g., DATE and TIME). Other such attributes can include, but are not limited to, the type of imager used, scanning information, digitizing information and the like. Alternatively, data attributes can be input by the subject and/or operator, for example subject
- 10 identifiers, i.e. characteristics associated with a particular subject. These identifiers include but are not limited to the following: (1) a subject code (e.g., a numeric or alpha-numeric sequence); (2) demographic information such as race, gender and age; (3) physical characteristics such as weight, height and body mass index (BMI); (4) selected aspects of the subject's medical history (e.g.,
- 15 disease states or conditions, etc.); and (5) disease-associated characteristics such as the type of bone disorder, if any; the type of medication used by the subject. In the practice of the present invention, each data point would typically be identified with the particular subject, as well as the demographic, etc. characteristic of that subject.
- 20 **[0120]** Other data attributes will be apparent to persons of ordinary skill in the art in light of the teachings of the present specification. (See, also, WO 02/30283, incorporated by reference in its entirety herein).

- 25 **[0121]** Thus, data (e.g., bone structural information or bone mineral density information or articular information) is obtained from normal control subjects using the methods described herein. These databases are typically referred to as "reference databases" and can be used to aid analysis of any given subject's image, for example, by comparing the information obtained from the subject to the reference database. Generally, the information obtained from the normal

control subjects will be averaged or otherwise statistically manipulated to provide a range of "normal" measurements. Suitable statistical manipulations and/or evaluations will be apparent to those of skill in the art in view of the teachings herein. The comparison of the subject's information to the reference database
5 can be used to determine if the subject's bone information falls outside the normal range found in the reference database or is statistically significantly different from a normal control.

[0122] Data obtained from images, as described above, can be manipulated, for example, using a variety of statistical analyses to produce useful information.
10 Databases can be created or generated from the data collected for an individual, or for a group of individuals, over a defined period of time (e.g., days, months or years), from derived data, and from data attributes.

[0123] For example, data can be aggregated, sorted, selected, sifted, clustered and segregated by means of the attributes associated with the data
15 points. A number of data mining software exist which can be used to perform the desired manipulations.

[0124] Relationships in various data can be directly queried and/or the data analyzed by statistical methods to evaluate the information obtained from manipulating the database.

20 **[0125]** For example, a distribution curve can be established for a selected data set, and the mean, median and mode calculated therefor. Further, data spread characteristics, e.g., variability, quartiles, and standard deviations can be calculated.

[0126] The nature of the relationship between any variables of interest can be
25 examined by calculating correlation coefficients. Useful methods for doing so include, but are not limited to: Pearson Product Moment Correlation and Spearman Rank Correlation. Analysis of variance permits testing of differences

among sample groups to determine whether a selected variable has a discernible effect on the parameter being measured.

[0127] Non-parametric tests can be used as a means of testing whether variations between empirical data and experimental expectancies are attributable to chance or to the variable or variables being examined. These include the Chi Square test, the Chi Square Goodness of Fit, the 2x2 Contingency Table, the Sign Test and the Phi Correlation Coefficient. Other tests include z-scores, T-scores or lifetime risk for arthritis, cartilage loss or osteoporotic fracture.

[0128] There are numerous tools and analyses available in standard data mining software that can be applied to the analyses of the databases that can be created according to this invention. Such tools and analysis include, but are not limited to, cluster analysis, factor analysis, decision trees, neural networks, rule induction, data driven modeling, and data visualization. Some of the more complex methods of data mining techniques are used to discover relationships that are more empirical and data-driven, as opposed to theory driven, relationships.

[0129] Statistical significance can be readily determined by those of skill in the art. The use of reference databases in the analysis of images facilitates that diagnosis, treatment and monitoring of bone conditions such as osteoporosis.

[0130] For a general discussion of statistical methods applied to data analysis, see Applied Statistics for Science and Industry, by A. Romano, 1977, Allyn and Bacon, publisher.

[0131] The data is preferably stored and manipulated using one or more computer programs or computer systems. These systems will typically have data storage capability (e.g., disk drives, tape storage, optical disks, etc.). Further, the computer systems can be networked or can be stand-alone systems. If networked, the computer system would be able to transfer data to any device

connected to the networked computer system for example a medical doctor or medical care facility using standard e-mail software, a central database using database query and update software (e.g., a data warehouse of data points, derived data, and data attributes obtained from a large number of subjects).

- 5 Alternatively, a user could access from a doctor's office or medical facility, using any computer system with Internet access, to review historical data that can be useful for determining treatment.

- [0132]** If the networked computer system includes a World Wide Web application, the application includes the executable code required to generate
10 database language statements, for example, SQL statements. Such executables typically include embedded SQL statements. The application further includes a configuration file that contains pointers and addresses to the various software entities that are located on the database server in addition to the different external and internal databases that are accessed in response to a user request.
15 The configuration file also directs requests for database server resources to the appropriate hardware, as can be necessary if the database server is distributed over two or more different computers.

- [0133]** As a person of skill in the art will appreciate, one or more of the parameters specified in Table 1, Table and Table 3 can be used at an initial time
20 point T_1 to assess the severity of a bone disease such as osteoporosis or arthritis. The patient can then serve as their own control at a later time point T_2 , when a subsequent measurement using one or more of the same parameters used at T_1 is repeated.

- [0134]** A variety of data comparisons can be made that will facilitate drug
25 discovery, efficacy, dosing, and comparisons. For example, one or more of the parameters specified in Table 1, Table 2 and Table 3 may be used to identify lead compounds during drug discovery. For example, different compounds can be tested in animal studies and the lead compounds with regard to highest

therapeutic efficacy and lowest toxicity, e.g. to the bone or the cartilage, can be identified. Similar studies can be performed in human subjects, e.g. FDA phase I, II or III trials. Alternatively, or in addition, one or more of the parameters specified in Table 1, Table 2 and Table 3 can be used to establish optimal dosing of a new compound. It will be appreciated also that one or more of the parameters specified in Table 1, Table 2 and Table 3 can be used to compare a new drug against one or more established drugs or a placebo. The patient can then serve as their own control at a later time point T_2 ,

EXAMPLES

10 **Example 1: Correlation of Macro-anatomical and Structural Parameters to Fracture Load**

[0135] Using 15 fresh cadaveric femurs, the following analyses were performed to determine the correlation of macro-anatomical and structural parameters to fracture load.

15 **[0136]** Standardization of Hip radiographs: Density and magnification calibration on the x-ray radiographs was achieved using a calibration phantom. The reference orientation of the hip x-rays was the average orientation of the femoral shaft.

20 **[0137]** Automatic Placement of Regions of Interest. An algorithm was developed and used to consistently and accurately place 7 regions of interest based on the geometric and position of proximal femur. Fig. 10. In brief, the algorithm involved the detection of femoral boundaries, estimation of shaft and neck axes, and construction of ROI based on axes and boundary intercept points. This approach ensured that the size and shape of ROIs placed conformed to the scale and shape of the femur, and thus were consistent relative to anatomic features on the femur.

25

[0138] Automatic Segmentation of the proximal femur. A global gray level thresholding using bi-modal histogram segmentation algorithm(s) was performed on the hip images and a binary image of the proximal femur was generated. Edge-detection analysis was also performed on the hip x-rays, including edge detection of the outline of the proximal femur that involved breaking edges detected into segments and characterizing the orientation of each segment. Each edge segment was then referenced to a map of expected proximal femur edge orientation and to a map of the probability of edge location. Edge segments that did not conform to the expected orientation or which were in low probability regions were removed. Morphology operations were applied to the edge image(s) to connect any discontinuities. The edge image formed an enclosed boundary of the proximal femur. The region within the boundary was then combined with the binary image from global thresholding to form the final mask of the proximal femur.

[0139] Automatic Segmentation and Measurement of the Femoral Cortex: Within a region of interest (ROI), edge detection was applied. Morphology operations were applied to connect edge discontinuities. Segments were formed within enclosed edges. The area and the major axis length of each segment were then measured. The regions were also superimposed on the original gray level image and average gray level within each region was measured. The cortex was identified as those segments connected to the boundary of the proximal femur mask with the greatest area, longest major axis length and a mean gray level about the average gray level of all enclosed segments within the proximal femur mask.

[0140] The segment identified as cortex was then skeletonized. The orientation of the cortex skeleton was verified to conform to the orientation map of the proximal femur edge. Euclidean distance transform was applied to the binary image of the segment. The values of distance transform value along the

skeleton were sampled and their average, standard deviation, minimum, maximum and mod determined.

[0141] *Watershed Segmentation for Characterizing Trabecular Structure:*

Marrow spacing was characterized by determining watershed segmentation of gray level trabecular structures on the hip images; essentially as described in Russ "The Image Processing Handbook," 3rd. ed. pp.494-501. This analysis take the gray level contrast between the marrow spacing and adjacent trabecular structures into account. The segments of marrow spacing generated using watershed segmentation were measured for the area, eccentricity, orientation, and the average gray level on the x-ray image within the segment. Mean, standard deviation, minimum, maximum and mod. were determined for each segment. In addition, various structural and/or macro-anatomical parameters were assessed for several ROIs (FIG. 10).

[0142] *Measurement of Femoral Neck BMD:* DXA analysis of bone mineral

density was performed in the femoral neck region of the femurs.

[0143] *Biomechanical Testing of Intact Femur:* Each cadaveric femur sample

(n=15) was tested for fracture load as follows. First, the femur was placed at a 15° angle of tilt and an 8° external rotation in an Instron 1331 Instrument (Instron, Inc.) and a load vector at the femoral head simulating single-leg stance was generated, essentially as described in Cheal et al. (1992) *J. Orthop. Res.* 10(3):405-422. Second, varus/ valgus and torsional resistive movements simulating passive knee ligaments restraints were applied. Next, forces and movement at failure were measured using a six-degree of freedom load cell. Subsequently, a single ramp, axial compressive load was applied to the femoral head of each sample at 100 mm/s until fracture. (FIG. 12). Fracture load and resultant equilibrium forces and moments at the distal end of the femur were measured continuously. FIG. 11 shows various results of biomechanical testing.

[0144] The correlation between (1) DXA femoral next BMD and fracture load; (2) bone structure and fracture load; and (3) macro-anatomical analyses and fracture load was determined and shown in FIG. 13-15, respectively.

[0145] Multivariate linear regression analysis was also performed, combining
5 several structural and macro-anatomical parameters, including local maximum marrow spacing ($r=0.6$ linearized); standard deviation of cortical thickness of ROI3 ($r=0.57$); maximum cortical thickness of ROI5 ($r=0.56$); and mean node-free end length for ROI3 ($r=0.50$). Results are shown in FIG. 16 and demonstrate that, using analyses, described herein there is a good correlation
10 between predicted fracture load and actual fracture load ($r=0.81$, $p<0.001$). The mean fracture load was 5.4 kiloNewton with a standard deviation of 2.3 kiloNewton. These statistics and the coefficients of multivariate linear regression were stored as data of the fracture load reference database.

Example 2: Correlation of 2D and 3D measurements

[0146] To demonstrate that methods using 2D x-ray technology to
15 quantitatively assess trabecular architecture is as effective as 3D μ CT, which serves as a gold standard for such measurements, the following experiments were performed. Bone cores ($n=48$) were harvested from cadaveric proximal femora. Specimen radiographs were obtained and 2D structural parameters were
20 measured on the radiographs. Cores were then subjected to 3D μ CT and biomechanical testing. The μ CT images were analyzed to obtain 3D micro-structural measurements. Digitized 2D x-ray images of these cores were also analyzed as described herein to obtain comparative micro-structural measurements.

[0147] Results showed very good correlation among the numerous 2D
25 parameters and 3D μ CT measurements, including for example correlation between 2D Trabecular Perimeter/Trabecular Area (Tb.P/Tb.A) with 3D Bone Surface/Bone Volume ($r = 0.92$, $p < 0.001$), and 2D Trabecular Separation

(Tb.Sp) with 3D Trabecular Separation ($r = 0.88$, $p < 0.001$). The 2D Tb.P/Tb.A and 2D Tb.Sp also function correlate very well as predictive parameters for the mechanical loads required to fracture the cores, with $r = -0.84$ ($p < 0.001$) and $r = -0.83$ ($p < 0.001$), respectively, when logarithmic and exponential transformations
5 were used in the regression.

[0148] These results demonstrate that 2D micro-structural measurements of trabecular bone from digitized radiographs are highly correlated with 3D measurements obtained from μ -CT images. Therefore, the mechanical characteristics of trabecular bone microstructure from digitized radiographic
10 images can be accurately determined from 2D images.

Example 3. Prediction of fracture risk using fracture load reference database.

[0149] A hip x-ray of cadaver pelvis was exposed using standard clinical procedure and equipment. The radiograph film was developed and digitized. The
15 image was then analyzed to obtain micro-structure, and macro-anatomical parameters. The local maximum spacing, standard deviation of cortical thickness of ROI3, maximum cortical thickness of ROI5, and mean node-free end length for ROI3 were used to predict load required to fracture the cadaver hip using the coefficients of multivariate linear regression stored in the fracture load reference
20 database. The predicted fracture load was 7.5 kiloNewton. This fracture load is 0.98 standard deviation above the average of the fracture load reference database (or z-score = 0.98). This result may suggest that the subject had a relatively low risk of sustaining a hip fracture as compared to the population of the reference database.

[0150] The foregoing description of embodiments of the present invention has
25 been provided for the purposes of illustration and description. It is not intended to be exhaustive or to limit the invention to the precise forms disclosed. Many modifications and variations will be apparent to the practitioner skilled in the art.

The embodiments were chosen and described in order to best explain the principles of the invention and its practical application, thereby enabling others skilled in the art to understand the invention and the various embodiments and with various modifications that are suited to the particular use contemplated. It is
5 intended that the scope of the invention be defined by the following claims and its equivalence.